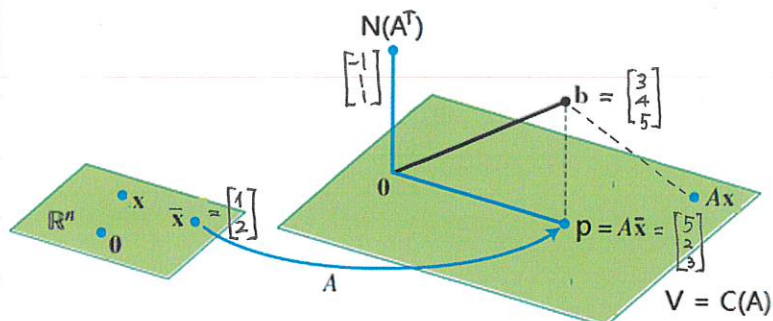


# **Chapter 4   Projections and Linear Transformations**

# Projection & Least square solution

$$\begin{cases} V = C(A) \triangleleft \mathbb{R}^m & \mathbb{R}^m = C(A) \oplus N(A^T) \\ b \in \mathbb{R}^m & b = p + (b-p) \end{cases}$$



•  $A_{m \times n} x = b$

(a)  $b \in C(A)$  有解,  $b = \begin{bmatrix} 5 \\ 2 \\ 3 \end{bmatrix}$ ,  $\begin{bmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 5 \\ 2 \\ 3 \end{bmatrix}$ ,  $\begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$

$A\bar{x} = p$  有解

(b)  $b \notin C(A)$  無解,  $b = \begin{bmatrix} 3 \\ 4 \\ 5 \end{bmatrix}$ ,  $\begin{bmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 3 \\ 4 \\ 5 \end{bmatrix}$ ,  $p = \text{proj}_V b = \begin{bmatrix} 5 \\ 2 \\ 3 \end{bmatrix}$ , LS 解  $\bar{x} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$

• 定理  $\bar{x}$ :  $Ax = b$  的 LS 解 ( $A\bar{x} = p$ )

(1)  $\|b - A\bar{x}\| \leq \|b - Ax\|, \forall x \in \mathbb{R}^n$  (最佳近似解)

(2)  $A^T A \bar{x} = A^T b$  (Normal Eq.)

( $\because b - A\bar{x} \in N(A^T) \Rightarrow A^T(b - A\bar{x}) = 0$ )

(3)  $\bar{x} = (A^T A)^{-1} A^T b$

$b = \begin{bmatrix} 3 \\ 4 \\ 5 \end{bmatrix}, A = \begin{bmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 1 \end{bmatrix} \Rightarrow \bar{x} = (A^T A)^{-1} A^T b = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$

(4)  $p = \text{proj}_V b = A(A^T A)^{-1} A^T b$

(5) Projection matrix  $P_V = A(A^T A)^{-1} A^T$

( $P_V^2 = P_V$ ) ( $V = \text{Span}(v_1, \dots, v_k), A = \begin{bmatrix} | & & | \\ v_1 & \dots & v_k \\ | & & | \end{bmatrix}$ )

• Lemma  $\text{rank}(A) = n \Rightarrow A^T A$  invertible

( $\because A^T A x = 0 \Rightarrow Ax \cdot Ax = x \cdot A^T A x = 0$   
 $\Rightarrow Ax = 0 \Rightarrow \underline{x = 0}$ )

• 討論  $P_V = A(A^T A)^{-1} A^T$

(1)  $b \in C(A) : P_V b = b$

$\because b = Ax$

(2)  $b \in N(A^T) : P_V b = 0$

$A^T b = 0$

(3)  $V = \text{Span}(a) : P_V b = \frac{1}{\|a\|^2} (a a^T) b = \frac{a \cdot b}{\|a\|^2} a$

$A = a$

(4)  $V = \{x \in \mathbb{R}^3 \mid x - 2y + z = 0\} = \text{Span}(\underbrace{\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}}_u, \underbrace{\begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}}_v)$

$A = \begin{bmatrix} 1 & -1 \\ 1 & 0 \\ 1 & 1 \end{bmatrix}$

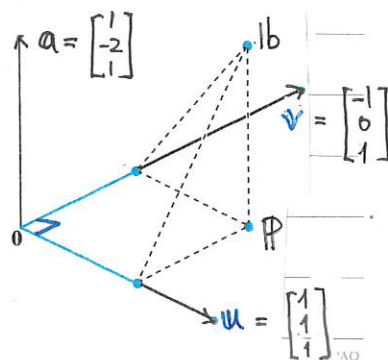
(甲)  $P_V = A(A^T A)^{-1} A^T = \frac{1}{6} \begin{bmatrix} 5 & 2 & -1 \\ 2 & 2 & 2 \\ -1 & 2 & 5 \end{bmatrix}$

(乙)  $P_V = I_3 - \frac{1}{\|a\|^2} a a^T = \frac{1}{6} \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}$

(丙)  $P_V = \begin{bmatrix} | & | \\ u & v \\ | & | \end{bmatrix} \left( \begin{bmatrix} | & | \\ u^T & v^T \\ | & | \end{bmatrix} \right)^{-1} \begin{bmatrix} | & | \\ u^T & v^T \\ | & | \end{bmatrix}$  ( $u \perp v$ )

$= \frac{u u^T}{\|u\|^2} + \frac{v v^T}{\|v\|^2} = \frac{1}{6} \begin{bmatrix} & & \\ & & \\ & & \end{bmatrix}$

$P_V b = \frac{b \cdot u}{\|u\|^2} u + \frac{b \cdot v}{\|v\|^2} v$  (定理 3)

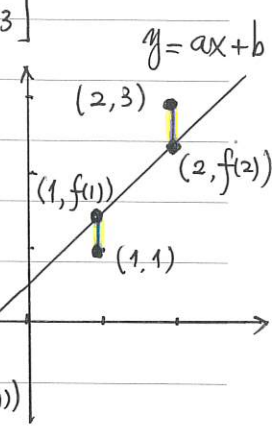


# Data fitting (Least square, Regression)

Example 1 求  $y = f(x) = ax + b$  thru  $(-1, 1), (1, 5), (2, 7)$

解 
$$\begin{cases} f(-1) = -a + b = 1 \\ f(1) = a + b = 5 \\ f(2) = 2a + b = 7 \end{cases} \Rightarrow \begin{bmatrix} f(-1) \\ f(1) \\ f(2) \end{bmatrix} = \begin{bmatrix} -1 & 1 \\ 1 & 1 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} 1 \\ 5 \\ 7 \end{bmatrix} \Rightarrow \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} 2 \\ 3 \end{bmatrix}$$

$Ax = b$



Example 2 求  $y = f(x) = ax + b$  thru best fits  $\begin{cases} (-1, 0) \\ (1, 1) \\ (2, 3) \end{cases}$

解 (1) 
$$\begin{bmatrix} f(-1) \\ f(1) \\ f(2) \end{bmatrix} = \begin{bmatrix} -1 & 1 \\ 1 & 1 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 3 \end{bmatrix}$$
 無解

$Ax = b$

最小平方 (迴歸) 直線  
LS (Regression) line

(2) 求 LS 解

$$\begin{aligned} \text{Min } D &= [f(-1) - 0]^2 + [f(1) - 1]^2 + [f(2) - 3]^2 = \left\| \begin{bmatrix} f(-1) \\ f(1) \\ f(2) \end{bmatrix} - \begin{bmatrix} 0 \\ 1 \\ 3 \end{bmatrix} \right\|^2 = \|Ax - b\|^2 \\ x = \begin{bmatrix} a \\ b \end{bmatrix} &= [(-a+b)-0]^2 + [(a+b)-1]^2 + [(2a+b)-3]^2 \end{aligned}$$

(甲) 微積分:

$$\begin{cases} \frac{\partial D}{\partial a} = 2[(-a+b-0)(-1)] + 2[(a+b-1)(1)] + 2[(2a+b-3)(2)] = 0 \\ \frac{\partial D}{\partial b} = 2[(-a+b-0)(1)] + 2[(a+b-1)(1)] + 2[(2a+b-3)(1)] = 0 \end{cases}$$

$$\begin{bmatrix} -1 & 1 & 2 \\ 1 & 1 & 1 \end{bmatrix} \begin{pmatrix} -a+b \\ a+b \\ 2a+b \end{pmatrix} - \begin{bmatrix} 0 \\ 1 \\ 3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

(2) 線性代數:  $A^T (Ax - b) = 0$

$$\therefore \begin{bmatrix} \bar{a} \\ \bar{b} \end{bmatrix} = \bar{x} = (A^T A)^{-1} A^T b = \frac{1}{14} \begin{bmatrix} 3 & -2 \\ -2 & 6 \end{bmatrix} \begin{bmatrix} 7 \\ 4 \end{bmatrix} = \frac{1}{14} \begin{bmatrix} 13 \\ 10 \end{bmatrix}$$

$$y = \frac{13}{14}x + \frac{10}{14}$$

$$\begin{cases} (1, -1) = (x_1, y_1) \\ (2, 4) = (x_2, y_2) \\ (3, 5) \\ (4, 7) \\ (5, 11) = (x_5, y_5) \end{cases}$$

Example 3 求拋物線  $y = f(x) = ax^2 + bx + c$  best fits

解  $\begin{cases} f(1) = a + b + c = -1 \\ f(2) = 4a + 2b + c = 4 \\ f(3) = 9a + 3b + c = 5 \\ f(4) = 16a + 4b + c = 7 \\ f(5) = 25a + 5b + c = 11 \end{cases} \Rightarrow \begin{bmatrix} f(1) \\ f(2) \\ f(3) \\ f(4) \\ f(5) \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 \\ 4 & 2 & 1 \\ 9 & 3 & 1 \\ 16 & 4 & 1 \\ 25 & 5 & 1 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} -1 \\ 4 \\ 5 \\ 7 \\ 11 \end{bmatrix}$

LS 解

$$A \bar{x} = \bar{b}$$

$$\begin{aligned} \text{Min } D &= [f(1)-1]^2 + [f(2)-4]^2 + [f(3)-5]^2 + [f(4)-7]^2 + [f(5)-11]^2 = \left\| \begin{bmatrix} f(1) \\ f(2) \\ f(3) \\ f(4) \\ f(5) \end{bmatrix} - \begin{bmatrix} -1 \\ 4 \\ 5 \\ 7 \\ 11 \end{bmatrix} \right\|^2 \\ \bar{x} = [a \ b \ c] &= \sum_{i=1}^5 [f(x_i) - y_i]^2 = \sum_{i=1}^5 [(ax_i^2 + bx_i + c) - y_i]^2 \end{aligned}$$

微積分

$$\begin{cases} \frac{\partial D}{\partial a} = \sum_{i=1}^5 2 [f(x_i) - y_i] x_i^2 = 0 \\ \frac{\partial D}{\partial b} = \sum_{i=1}^5 2 [f(x_i) - y_i] x_i = 0 \\ \frac{\partial D}{\partial c} = \sum_{i=1}^5 2 [f(x_i) - y_i] \cdot 1 = 0 \end{cases} \Rightarrow \begin{bmatrix} x_1^2 & x_2^2 & x_3^2 & x_4^2 & x_5^2 \\ x_1 & x_2 & x_3 & x_4 & x_5 \\ 1 & 1 & 1 & 1 & 1 \end{bmatrix} \begin{pmatrix} f(x_1) \\ f(x_2) \\ f(x_3) \\ f(x_4) \\ f(x_5) \end{pmatrix} - \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \\ y_5 \end{pmatrix} = 0$$

線性代數

$$A^T (A \bar{x} - \bar{b}) = 0$$

Example 4  $y = f(x) = a + bx + c \sin x + d e^x$  best fits

$$\begin{cases} (1, -1) \\ (2, 4) \\ (3, 5) \\ (4, 7) \\ (5, 11) \end{cases}$$

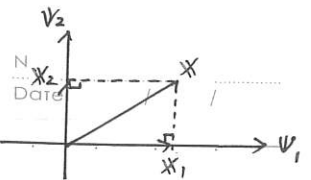
$$\begin{bmatrix} f(x_1) \\ f(x_2) \\ f(x_3) \\ f(x_4) \\ f(x_5) \end{bmatrix} = \begin{bmatrix} 1 & 1 & \sin 1 & e^1 \\ 1 & 2 & \sin 2 & e^2 \\ 1 & 3 & \sin 3 & e^3 \\ 1 & 4 & \sin 4 & e^4 \\ 1 & 5 & \sin 5 & e^5 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} = \begin{bmatrix} -1 \\ 4 \\ 5 \\ 7 \\ 11 \end{bmatrix}$$

thru:  $A \bar{x} = \bar{b}$   
best fits:  $A^T A \bar{x} = A^T \bar{b}$



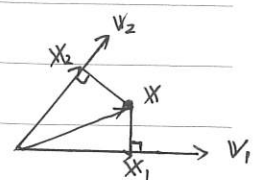
# Orthogonal Bases

正交基底



定義  $\{v_1, \dots, v_n\}$  orthogonal :  $v_i \cdot v_j = 0 \quad i \neq j$   
 $\{ \quad \quad \}$  orthonormal :  $v_i \cdot v_j = \delta_{ij} = \begin{cases} 0 & i \neq j \\ 1 & i = j \end{cases}$

$x = x_1 + x_2$   
 $x \neq x_1 + x_2$



定理1'  $\{u, v, w\}$  orthogonal  $\Rightarrow$  independent

証  $au + bv + cw = 0 \begin{cases} \cdot u & \Rightarrow a = 0 \\ \cdot v & \Rightarrow b = 0 \\ \cdot w & \Rightarrow c = 0 \end{cases}$

定理2'  $\{u, v, w\}$  basis for  $V$ , 則  $\{u, v, w\}$  orthogonal  $\Leftrightarrow \forall x \in V, x = \frac{x \cdot u}{\|u\|^2} u + \frac{x \cdot v}{\|v\|^2} v + \frac{x \cdot w}{\|w\|^2} w$

証 " $\Rightarrow$ "  $x = au + bv + cw, x \cdot u = a \|u\|^2$

" $\Leftarrow$ " (取  $x = u$ )  $u = \frac{u \cdot u}{\|u\|^2} u + \frac{u \cdot v}{\|v\|^2} v + \frac{u \cdot w}{\|w\|^2} w \Rightarrow \begin{cases} u \cdot v = 0 \\ u \cdot w = 0 \end{cases}$   
 (唯一性)  $= 1u + 0v + 0w$

(dim  $V = 2$ )

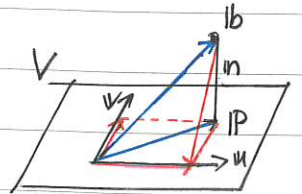
定理3'  $V \subset \mathbb{R}^3, \{u, v\}$  orthogonal basis  $\Leftrightarrow \forall b \in \mathbb{R}^3, \text{proj}_V b = P = \frac{b \cdot u}{\|u\|^2} u + \frac{b \cdot v}{\|v\|^2} v$

証 " $\Rightarrow$ "  $P = \frac{P \cdot u}{\|u\|^2} u + \frac{P \cdot v}{\|v\|^2} v,$

$= \frac{b \cdot u}{\|u\|^2} u + \frac{b \cdot v}{\|v\|^2} v \quad (\because b = P + n \Rightarrow b \cdot u = P \cdot u)$

或=垂線定理

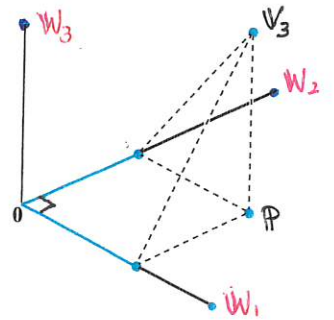
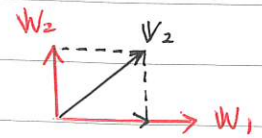
" $\Leftarrow$ " (定理2')  $\forall b \in V$



定理4' (Gram-Schmidt process)  $\{v_1, v_2, v_3\}$  basis  $\xrightarrow{\text{正交化}}$   $\{w_1, w_2, w_3\}$  orthogonal basis  $\xrightarrow{q_i = \frac{w_i}{\|w_i\|}}$   $\{q_1, q_2, q_3\}$  orthonormal basis

Process  $\begin{cases} (1) w_1 = v_1 \\ (2) w_2 = v_2 - \frac{v_2 \cdot w_1}{\|w_1\|^2} w_1 \\ (3) w_3 = v_3 - \frac{v_3 \cdot w_1}{\|w_1\|^2} w_1 - \frac{v_3 \cdot w_2}{\|w_2\|^2} w_2 \end{cases}$

$\Rightarrow \begin{cases} (1) \{w_1, w_2, w_3\}$  orthogonal  $P$  \\ (2)  $\text{Span}(w_1, w_2, w_3) = \text{Span}(v_1, v_2, v_3)$  \end{cases}



Example  $v_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}, v_2 = \begin{bmatrix} 3 \\ 1 \\ -1 \\ 1 \end{bmatrix}, v_3 = \begin{bmatrix} 1 \\ 1 \\ 3 \\ 3 \end{bmatrix}$

$\begin{cases} w_1 = v_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} \\ w_2 = v_2 - \frac{v_2 \cdot w_1}{\|w_1\|^2} w_1 = v_2 - w_1 = \begin{bmatrix} 2 \\ 0 \\ -2 \\ 0 \end{bmatrix} \\ w_3 = v_3 - \frac{v_3 \cdot w_1}{\|w_1\|^2} w_1 - \frac{v_3 \cdot w_2}{\|w_2\|^2} w_2 = v_3 - 2w_1 + \frac{1}{2} w_2 = \begin{bmatrix} 0 \\ -1 \\ 0 \\ 1 \end{bmatrix} \end{cases}$

$\begin{cases} q_1 = \frac{1}{2} w_1 = \frac{1}{2} \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} \\ q_2 = \frac{1}{2\sqrt{2}} w_2 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ -1 \\ 0 \end{bmatrix} \\ q_3 = \frac{1}{\sqrt{2}} w_3 = \frac{1}{\sqrt{2}} \begin{bmatrix} 0 \\ -1 \\ 0 \\ 1 \end{bmatrix} \end{cases}$

# Orthogonal Bases

No. \_\_\_\_\_  
Date: / /

定理 1  $\{v_1, \dots, v_k\}$  orthogonal  $\Rightarrow$  independent

定理 2  $\{v_1, \dots, v_k\}$  basis for  $V$ , 則  $\{v_1, \dots, v_k\}$  orthogonal  $\Leftrightarrow \forall x \in V, x = \sum_{i=1}^k \frac{x \cdot v_i}{\|v_i\|^2} v_i$

定理 3  $V \triangleleft \mathbb{R}^n, \dim V = k$ , 則  $\{v_1, \dots, v_k\}$  orthogonal basis  $\Leftrightarrow \forall b \in \mathbb{R}^n, \text{proj}_V b = \sum_{i=1}^k \frac{b \cdot v_i}{\|v_i\|^2} v_i$

定理 4 (Gram-Schmidt Process) basis  $\{v_1, \dots, v_k\}$   $\xrightarrow{\text{正交化}}$   $\{w_1, \dots, w_k\}$   $\xrightarrow{\text{單位長}}$   $\{q_1, \dots, q_k\}$   
orthogonal  orthonormal

Process

$$\left\{ \begin{array}{l} (1) w_1 = v_1 \\ (2) w_2 = v_2 - \frac{v_2 \cdot w_1}{\|w_1\|^2} w_1 \\ (3) w_3 = v_3 - \frac{v_3 \cdot w_1}{\|w_1\|^2} w_1 - \frac{v_3 \cdot w_2}{\|w_2\|^2} w_2 \\ \vdots \\ (k) w_k = v_k - \frac{v_k \cdot w_1}{\|w_1\|^2} w_1 - \frac{v_k \cdot w_2}{\|w_2\|^2} w_2 - \dots - \frac{v_k \cdot w_{k-1}}{\|w_{k-1}\|^2} w_{k-1} \end{array} \right. \Rightarrow \left\{ \begin{array}{l} (1) \{w_1, \dots, w_k\} \text{ orthogonal} \\ (2) \text{Span}(w_1, \dots, w_k) = \text{Span}(v_1, \dots, v_k) \end{array} \right.$$

## QR-decomposition

$$\left\{ \begin{array}{l} v_1 = w_1 = 2q_1 \\ v_2 = w_1 + w_2 = 2q_1 + 2\sqrt{2}q_2 \\ v_3 = 2w_1 - \frac{1}{2}w_2 + w_3 = 4q_1 - \sqrt{2}q_2 + \sqrt{2}q_3 \end{array} \right. \Rightarrow \begin{bmatrix} | & | & | \\ v_1 & v_2 & v_3 \\ | & | & | \end{bmatrix}_{4 \times 3} = \begin{bmatrix} | & | & | \\ q_1 & q_2 & q_3 \\ | & | & | \end{bmatrix}_{4 \times 3} \begin{bmatrix} 2 & 2 & 4 \\ 0 & 2\sqrt{2} & -\sqrt{2} \\ 0 & 0 & \sqrt{2} \end{bmatrix}_{3 \times 3}$$

$A = Q R$

• 定理 (QR-decomposition)

$$A_{m \times n} = Q_{m \times n} R_{n \times n} \quad \left\{ \begin{array}{l} Q^T Q = I_n \text{ (columns orthonormal)} \\ R: \text{upper triangular, nonsingular, 對角線} > 0 \end{array} \right.$$

• Remarks

(1) LS 解:  $\bar{x} = R^{-1} Q^T b \quad \because \bar{x} = (A^T A)^{-1} A^T b = (R^T Q^T Q R)^{-1} R^T Q^T b = R^{-1} Q^T b$

(2) Projection matrix  $P_V = Q Q^T \quad \because P_V = A(A^T A)^{-1} A^T = Q R R^{-1} Q^T$

$$= q_1 q_1^T + q_2 q_2^T + q_3 q_3^T \quad P_V b = (b \cdot q_1) q_1 + (b \cdot q_2) q_2 + (b \cdot q_3) q_3$$

$$= P_1 + P_2 + P_3$$



# Change of Basis

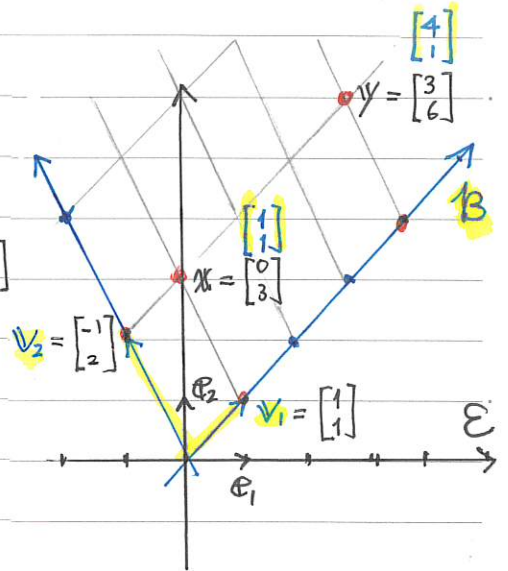
• 定理  $T: \mathbb{R}^n \rightarrow \mathbb{R}^n$ ,  $\mathbb{R}^n$ : 两基底  $\mathcal{E} = \{e_1, \dots, e_n\}$ ,  $\mathcal{B} = \{v_1, \dots, v_n\}$

(1) 点  $X$  在  $\mathcal{E}$  座標  $X = c_1 v_1 + \dots + c_n v_n$   
 在  $\mathcal{B}$  座標  $X' = C_{\mathcal{B}}(X) = \begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix}$   $\Rightarrow \begin{cases} X = P X' \\ X' = P^{-1} X \end{cases}$ ,  $P = \begin{bmatrix} | & | \\ v_1 & \dots & v_n \\ | & | \end{bmatrix}$   $\mathcal{B} \rightarrow \mathcal{E}$  座標變換矩陣

(2)  $\begin{cases} [T]_{\mathcal{E}} = A = \begin{bmatrix} | & | \\ T(e_1) & \dots & T(e_n) \\ | & | \end{bmatrix} \\ [T]_{\mathcal{B}} = A' = \begin{bmatrix} | & | \\ C_{\mathcal{B}}(T(v_1)) & \dots & C_{\mathcal{B}}(T(v_n)) \\ | & | \end{bmatrix} \end{cases} \Rightarrow \begin{cases} A = P A' P^{-1} \\ A' = P^{-1} A P \end{cases} \because \begin{cases} AX = Y \Rightarrow APX' = PY' \\ A'X' = Y' \end{cases}$

• Example  $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$

$\mathcal{E} = \{e_1, e_2\}$   $X = \begin{bmatrix} 0 \\ 3 \end{bmatrix} \xrightarrow{A = \begin{bmatrix} 3 & 1 \\ 2 & 2 \end{bmatrix}} Y = \begin{bmatrix} 3 \\ 6 \end{bmatrix}$   
 $\begin{bmatrix} 1 & -1 \\ 1 & 2 \end{bmatrix} = P$   $P^{-1} = \frac{1}{3} \begin{bmatrix} 2 & 1 \\ -1 & 1 \end{bmatrix}$   
 $\mathcal{B} = \{v_1, v_2\}$   $X' = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \xrightarrow{A' = \begin{bmatrix} 4 & 0 \\ 0 & 1 \end{bmatrix}} Y' = \begin{bmatrix} 4 \\ 1 \end{bmatrix}$



$A' = P^{-1} A P$   
 $\begin{bmatrix} 4 & 0 \\ 0 & 1 \end{bmatrix} = \frac{1}{3} \begin{bmatrix} 2 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 3 & 1 \\ 2 & 2 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ 1 & 2 \end{bmatrix}$

$Y' = \begin{bmatrix} 4 \\ 1 \end{bmatrix} \xleftarrow{v_2} \begin{bmatrix} 3 \\ 6 \end{bmatrix} \xleftarrow{v_1} \begin{bmatrix} 0 \\ 3 \end{bmatrix} \xleftarrow{e_1} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = X'$  ( $A \sim A'$ ,  $A$  is similar to  $A'$ )

• Remarks  $A = \begin{bmatrix} 3 & 1 \\ 2 & 2 \end{bmatrix}_{\mathcal{E}}$   $A' = \begin{bmatrix} 4 & 0 \\ 0 & 1 \end{bmatrix}_{\mathcal{B}} = P^{-1} A P$ ,  $P = \begin{bmatrix} 1 & -1 \\ 1 & 2 \end{bmatrix}$  ( $A$ : 可对角化)

(1)  $\begin{cases} T(v_1) = 4v_1, & \begin{bmatrix} 1 \\ 1 \end{bmatrix} \rightarrow \begin{bmatrix} 4 \\ 4 \end{bmatrix}, & \begin{bmatrix} 1 \\ 0 \end{bmatrix} \rightarrow \begin{bmatrix} 4 \\ 0 \end{bmatrix}, \\ T(v_2) = 1v_2, & \begin{bmatrix} -1 \\ 2 \end{bmatrix} \rightarrow \begin{bmatrix} -1 \\ 2 \end{bmatrix}, & \begin{bmatrix} 0 \\ 1 \end{bmatrix} \rightarrow \begin{bmatrix} 0 \\ 1 \end{bmatrix} \end{cases}$   $T(v) = Av = \lambda v$   $\begin{cases} \lambda: \text{eigen value of } A \\ v: \text{vector of } T \end{cases}$

(2)  $A = \begin{bmatrix} 3 & 1 \\ 2 & 2 \end{bmatrix}$ , 求  $\begin{cases} \lambda: (A - \lambda I)v = 0 \Rightarrow (A - \lambda I) \text{ singular} \Rightarrow \det(A - \lambda I) = 0 \\ v: \begin{cases} \lambda_1 = 4 \begin{bmatrix} -1 & 1 \\ 2 & -2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \Rightarrow v_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \\ \lambda_2 = 1 \begin{bmatrix} 2 & 1 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \Rightarrow v_2 = \begin{bmatrix} -1 \\ 2 \end{bmatrix} \end{cases} \end{cases}$  ( $\det \begin{bmatrix} 3-\lambda & 1 \\ 2 & 2-\lambda \end{bmatrix} = \lambda^2 - 5\lambda + 4$ )  
 $P = \begin{bmatrix} 1 & -1 \\ 1 & 2 \end{bmatrix}$ ,  $P^{-1} A P = \begin{bmatrix} 4 & 0 \\ 0 & 1 \end{bmatrix}$

(3)  $A = P \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix} P^{-1} \Rightarrow A^k = P \begin{bmatrix} \lambda_1 & \\ & \lambda_2 \end{bmatrix}^k P^{-1} = P \begin{bmatrix} \lambda_1^k & \\ & \lambda_2^k \end{bmatrix} P^{-1}$

(4)  $X_k = A X_{k-1} \Rightarrow X_k = A^2 X_{k-2} = \dots = A^k X_0 = P \begin{bmatrix} \lambda_1^k & \\ & \lambda_2^k \end{bmatrix} P^{-1} X_0 = [v_1 \ v_2] \begin{bmatrix} \lambda_1^k \\ \lambda_2^k \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix}$

$\begin{cases} a_k = 3a_{k-1} + b_{k-1}, \\ b_k = 2a_{k-1} + 2b_{k-1}, \end{cases} \quad k \geq 1, \quad \begin{cases} a_0 = 3 \\ b_0 = 9 \end{cases} \quad = \underline{c_1 \lambda_1^k v_1 + c_2 \lambda_2^k v_2}$

$\begin{bmatrix} a_k \\ b_k \end{bmatrix} = \begin{bmatrix} 3 & 1 \\ 2 & 2 \end{bmatrix} \begin{bmatrix} a_{k-1} \\ b_{k-1} \end{bmatrix}$ ,  $\begin{bmatrix} a_0 \\ b_0 \end{bmatrix} = \begin{bmatrix} 3 \\ 9 \end{bmatrix} \Rightarrow X_k = \begin{bmatrix} a_k \\ b_k \end{bmatrix} = \begin{bmatrix} 1 & -1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 4^k & 0 \\ 0 & 1^k \end{bmatrix} \begin{bmatrix} 5 \\ 2 \end{bmatrix} = \underline{5 \cdot 4^k \begin{bmatrix} 1 \\ 1 \end{bmatrix} + 2 \cdot 1^k \begin{bmatrix} -1 \\ 2 \end{bmatrix}}$   
 $X_k = A X_{k-1}$ ,  $X_0 = \begin{bmatrix} 3 \\ 9 \end{bmatrix}$

# 應用

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(1) 到  $V = \text{span}(v_1, v_2)$  的投影矩陣 (Example 1)

$$B = \{v_1, v_2, a\}, a \in V^\perp$$

$$P_V = P \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} P^{-1}$$

$$\varepsilon: \mathbb{R}^n \xrightarrow{A} \mathbb{R}^n$$

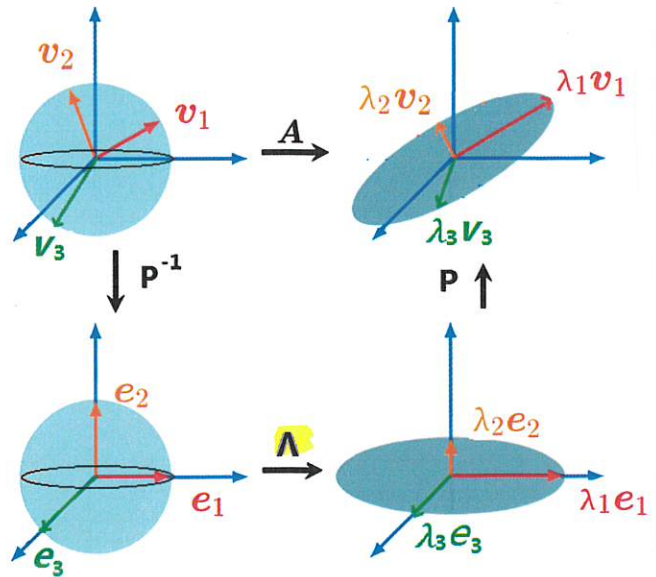
$$B: \mathbb{R}^n \xrightarrow{A'} \mathbb{R}^n$$

$$A = P A' P^{-1}$$

(2) 繞  $a$  旋轉  $\theta$  角 (Example 2)

$$B = \{v_1, v_2, a\}, v_1, v_2 \in V = a^\perp$$

$$R_\theta = P \begin{bmatrix} \cos\theta & -\sin\theta & 0 \\ \sin\theta & \cos\theta & 0 \\ 0 & 0 & 1 \end{bmatrix} P^{-1}$$



(3) Eigenvalues decomposition

$$B = \{v_1, \dots, v_n\} \quad v_i: \text{eigen vector}$$

$$A = P \begin{bmatrix} \lambda_1 & & \\ & \dots & \\ & & \lambda_n \end{bmatrix} P^{-1} \quad \begin{cases} A v_i = \lambda_i v_i \\ (\text{作用方向不變}) \end{cases}$$

(4) Orthogonal decomposition

$$B = \{q_1, \dots, q_n\} \text{ orthonormal eigen basis}$$

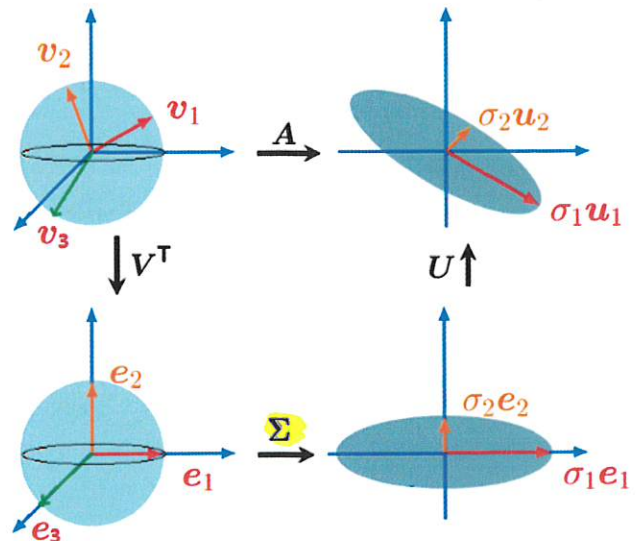
$$A = Q \begin{bmatrix} \lambda_1 & & \\ & \dots & \\ & & \lambda_n \end{bmatrix} Q^T \quad \begin{cases} A: \text{symmetric} \\ \Rightarrow \text{Quadratic form, Constraint optimization, PCA} \end{cases}$$

(5) Singular-values decomposition (SVD)

$U$ : orthonormal basis for  $\mathbb{R}^m$

$V$ : " " "  $\mathbb{R}^n$

$$A_{m \times n} = U \Sigma V^T \quad \begin{cases} v_i: \text{最大作用方向} \\ \Rightarrow \text{影像壓縮} \end{cases}$$



(6) Jordan forms

$$A = P \begin{bmatrix} \lambda_1 & 1 & & \\ & \lambda_1 & 1 & \\ & & \lambda_1 & 1 \\ & & & \lambda_2 & 1 \\ & & & & \lambda_2 \end{bmatrix} P^{-1}$$



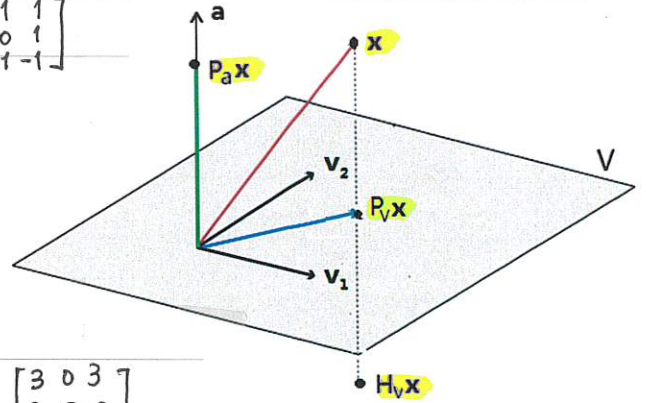
• Example 1  $V = \{x \in \mathbb{R}^3 \mid x - 2y - z = 0\}$ ,  $P_V, P_a, H_V = ?$   $H_V$ : 对  $V$  作 镜射.  
 $a = \begin{bmatrix} 1 \\ -2 \\ -1 \end{bmatrix}$ ,  $V = \text{span} \left( \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix} \right)$   
 $\begin{cases} P_V + P_a = I_3 \\ P_V - P_a = H_V \end{cases}$

(A) 基底  $\mathcal{E} = \{e_1, e_2, e_3\}$

(1)  $P_V = A(A^T A)^{-1} A^T = \frac{1}{6} \begin{bmatrix} 5 & 2 & 1 \\ 2 & 2 & -2 \\ 1 & -2 & 5 \end{bmatrix}$ ,  $A = \begin{bmatrix} 1 & 1 \\ 0 & 1 \\ 1 & -1 \end{bmatrix}$

(2)  $P_a = \frac{1}{\|a\|^2} a a^T = \frac{1}{6} \begin{bmatrix} 1 & -2 & -1 \\ -2 & 4 & 2 \\ -1 & 2 & 1 \end{bmatrix}$

(3)  $H_V = P_V - P_a = \frac{1}{6} \begin{bmatrix} 4 & 4 & 2 \\ 4 & -2 & -4 \\ 2 & -4 & 4 \end{bmatrix}$



(B) 基底  $\mathcal{B} = \{v_1, v_2, a\}$ ,  $P = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & -2 \\ 1 & -1 & -1 \end{bmatrix}$ ,  $P^{-1} = \frac{1}{6} \begin{bmatrix} 3 & 0 & 3 \\ 2 & 2 & -2 \\ 1 & -2 & -1 \end{bmatrix}$

(1)  $P'_V = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \Rightarrow P_V = P P'_V P^{-1} = \frac{1}{6} \begin{bmatrix} 5 & 2 & 1 \\ 2 & 2 & -2 \\ 1 & -2 & 5 \end{bmatrix}$

(2)  $P'_a = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \Rightarrow P_a = P P'_a P^{-1} = \frac{1}{6} \begin{bmatrix} 1 & -2 & -1 \\ -2 & 4 & 2 \\ -1 & 2 & 1 \end{bmatrix}$

(3)  $H'_V = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{bmatrix} \Rightarrow H_V = P H'_V P^{-1} = \frac{1}{6} \begin{bmatrix} 4 & 4 & 2 \\ 4 & -2 & -4 \\ 2 & -4 & 4 \end{bmatrix}$

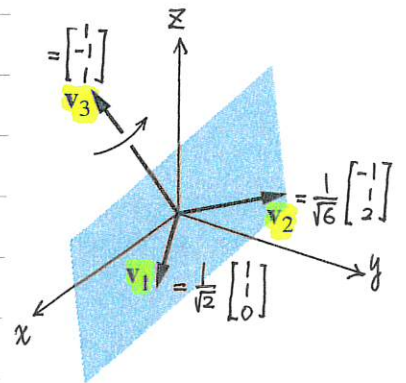
$$\begin{cases} P'_V(v_1) = v_1 \\ P'_V(v_2) = v_2 \\ P'_V(a) = 0 \end{cases}$$

• Example 2  $T: \mathbb{R}^3 \rightarrow \mathbb{R}^3$  rotate around  $\begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$  by  $\theta = \frac{2}{3}\pi$

$\mathcal{B} = \{v_1, v_2, v_3\}$ , rotate around  $v_3$  by  $\theta = \frac{2}{3}\pi$

$$A' = \begin{bmatrix} \cos\theta & -\sin\theta & 0 \\ \sin\theta & \cos\theta & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} -\frac{1}{2} & -\frac{\sqrt{3}}{2} & 0 \\ \frac{\sqrt{3}}{2} & -\frac{1}{2} & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$P = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{-1}{\sqrt{6}} & 1 \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{6}} & -1 \\ 0 & \frac{2}{\sqrt{6}} & 1 \end{bmatrix}$$



$$R_\theta = P A' P^{-1} = \begin{bmatrix} 0 & -1 & 0 \\ 0 & 0 & -1 \\ 1 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & -1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\begin{bmatrix} -y \\ -z \\ x \end{bmatrix} \leftarrow \begin{bmatrix} -y \\ x \\ z \end{bmatrix} \leftarrow \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

# Linear Transformations on Abstract Vector Spaces

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$\left\{ \begin{array}{l} V \\ W \end{array} \right.$  (abstract) vector spaces,  $T: V \rightarrow W$  線性變換:  $\begin{cases} T(u+v) = T(u) + T(v) \\ T(cu) = cT(u) \end{cases} \quad \forall \begin{array}{l} u, v \in V \\ c \in \mathbb{R} \end{array}$

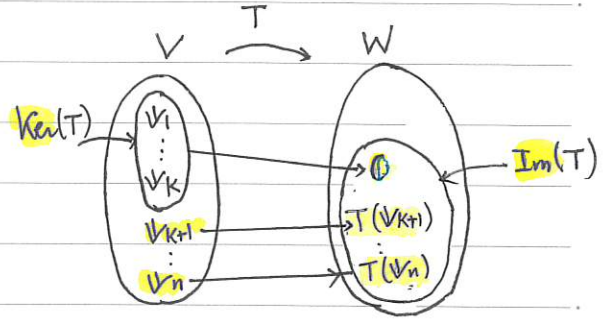
定理  $T \in L(V, W) = \{ \mathcal{T}: V \rightarrow W \mid \mathcal{T}: \text{linear} \}$   $\begin{cases} \dim V = n \\ \dim W = m \end{cases}$  基底  $V = \{v_1, \dots, v_n\}$   
 $W = \{w_1, \dots, w_m\}$

(1)  $\begin{cases} \text{Ker}(T) = \{x \in V \mid T(x) = 0\} \triangleleft V \\ \text{Im}(T) = \{T(x) \mid x \in V\} \triangleleft W \end{cases}$

(2)  $\dim \text{Ker}(T) + \dim \text{Im}(T) = n$

(3)  $\begin{cases} 0 \in \text{Ker}(T) \\ \text{Ker}(T) = \{0\} \iff T: 1-1 \end{cases}$

(4)  $\text{rank}(A) = m \implies T: \text{onto}$



証 (2) 設  $\dim \text{Ker}(T) = k$ , 基底為  $\{v_1, \dots, v_k\}$   $\Rightarrow \{T(v_{k+1}), \dots, T(v_n)\}$  基底  
擴充  $\{v_1, \dots, v_k\} \cup \{v_{k+1}, \dots, v_n\}$  為  $V$  之基底

(甲) (生成)  $\forall w \in \text{Im}(T), \exists v = \sum_{i=1}^n c_i v_i, w = T(v) = \sum_{i=1}^n c_i T(v_i)$   
(乙) (獨立)  $\sum_{i=k+1}^n c_i T(v_i) = 0 \Rightarrow T\left(\sum_{i=k+1}^n c_i v_i\right) = 0 \Rightarrow \sum_{i=k+1}^n c_i v_i \in \text{Ker}(T)$   
 $\Rightarrow \sum_{i=k+1}^n c_i v_i = \sum_{i=1}^k c_i v_i \Rightarrow c_i = 0, 1 \leq i \leq n$

$\mathcal{P}_k = \{a_0 + a_1 t + \dots + a_k t^k \mid a_i \in \mathbb{R}\}$

## Example

(a)  $D: \mathcal{P}_3 \rightarrow \mathcal{P}_2$   
 $f \mapsto f'$

$\text{Ker}(T)$	$\text{Im}(T)$
$\mathcal{P}_0$	$\mathcal{P}_2$

(b)  $M: \mathcal{C}^0([1, 2]) \rightarrow \mathcal{C}^0([1, 2])$   
 $f \mapsto tf$

$\{0\}$	$\mathcal{C}^0([1, 2])$
$tf(t) \equiv 0 \Rightarrow f(t) \equiv 0,$	$\frac{f(t)}{t} \mapsto f(t)$

(c)  $\mathcal{J}: \mathcal{P}_2 \rightarrow \mathbb{R}^2$   
 $f \mapsto [f(0) \ f(1)]$

$\{at(t-1) \mid a \in \mathbb{R}\}$	$\mathbb{R}^2$
	$a + (b-a)t \mapsto [a \ b]$

(d)  $\mathcal{S}: \mathcal{P}_2 \rightarrow \mathbb{R}^3$   
 $f \mapsto [f(0) \ f(1) \ f(2)]$

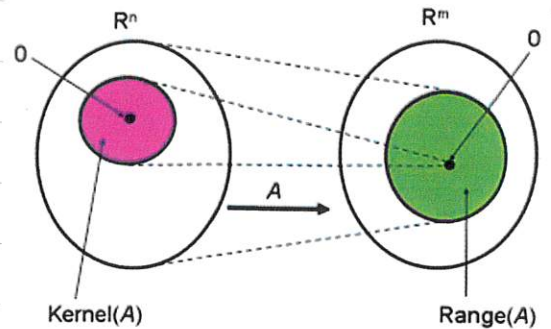
$\{0\}$	$\mathbb{R}^3$ (Lagrange Int.)
	$\frac{a}{2}(t-1)(t-2) - b t(t-2) + \frac{c}{2} t(t-1)$
	$\mapsto [a \ b \ c]$

(e)  $\mathcal{S}': \mathcal{P}_1 \rightarrow \mathbb{R}^3$   
 $f \mapsto [f(0) \ f(1) \ f(2)]$

$\{0\}$	$\{[a \ b \ c] \mid a-2b+c=0\}$
$\begin{cases} p = a \\ p+q = b \\ p+2q = c \end{cases} \iff$	$p+qt \mapsto [a \ b \ c]$

$$A_{m \times n}, T = \mathcal{M}_A : \mathbb{R}^n \rightarrow \mathbb{R}^m \quad \begin{cases} \text{Ker}(T) = N(A) \\ \text{Im}(T) = C(A) \end{cases}$$

- 定理 4.13
- $$\begin{cases} \dim N(A) + \dim C(A) = n \\ \dim \text{Ker}(T) + \dim \text{Im}(T) = n \end{cases}$$
- (1)  $\dim \text{Ker}(T) + \dim \text{Im}(T) = n$
- (2)  $\text{rank}(A) = \dim C(A) = \dim \text{Im}(T) \leq n$

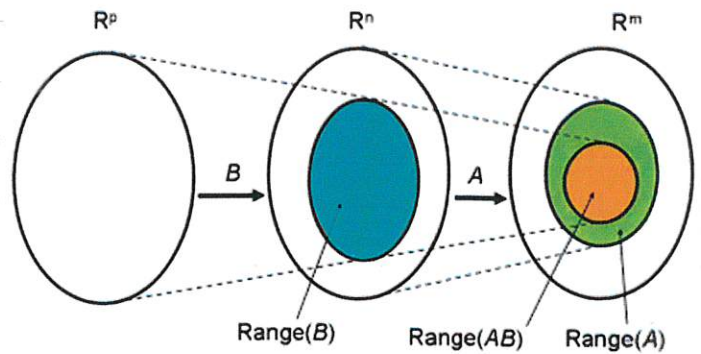


• 定理 4.14  $A_{m \times n}, B_{n \times p}$

- (1)  $\text{rank}(AB) \leq \text{rank}(A)$
- (2)  $n=p, B \text{ nonsingular} \Rightarrow \text{rank}(AB) = \text{rank}(A)$        $\text{rank}(B) = n, B: \text{满秩}$
- (3)  $\text{rank}(AB) \leq \text{rank}(B)$
- (4)  $m=n, A \text{ nonsingular} \Rightarrow \text{rank}(AB) = \text{rank}(B)$

証

- (1)  $\text{rank}(A) \geq \text{rank}(AB)$
- (2)  $\text{rank}(AB) \geq \text{rank}(AB\bar{B}^{-1}) = \text{rank}(A)$
- (3)  $A : \text{Im}(B) \rightarrow \mathbb{R}^m$
- 定理 4.13 (2)  $\Rightarrow$





# Change of Basis formula

• vector space  $V$ , 基底  $B = \{v_1, \dots, v_n\}$ ,  $x = x_1 v_1 + \dots + x_n v_n \in V$ ,  $C_B(x) = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \in \mathbb{R}^n$

•  $T: V \rightarrow W$  線性變換  $\begin{cases} V \text{ 基底 } \mathcal{V} = \{v_1, \dots, v_n\}, & \mathcal{V}' = \{v'_1, \dots, v'_n\} \\ W \text{ 基底 } \mathcal{W} = \{w_1, \dots, w_m\}, & \mathcal{W}' = \{w'_1, \dots, w'_m\} \end{cases}$

$$A = [T]_{\mathcal{V}, \mathcal{W}} = \begin{bmatrix} | & | & | \\ C_{\mathcal{W}}(T(v_1)) & \dots & C_{\mathcal{W}}(T(v_j)) & \dots & C_{\mathcal{W}}(T(v_n)) \\ | & | & | \end{bmatrix}, \begin{cases} T(v_j) = a_{1j} w_1 + \dots + a_{mj} w_m \\ a_j = C_{\mathcal{W}}(T(v_j)) = \begin{bmatrix} a_{1j} \\ \vdots \\ a_{mj} \end{bmatrix} \end{cases}$$

Example  $D: \mathcal{P}_3 \rightarrow \mathcal{P}_2$ ,  $\begin{cases} \mathcal{V} = \{1, t, t^2, t^3\}, & \mathcal{V}' = \{1, t-1, (t-1)^2, (t-1)^3\} \\ f \mapsto f' & \mathcal{W} = \{1, t, t^2\}, & \mathcal{W}' = \{1, t-1, (t-1)^2\} \end{cases}$

$$f(t) = 2 - t + 5t^2 + 4t^3 \xrightarrow{D} f'(t) = -1 + 10t + 12t^2$$

$$= 10 + 21(t-1) + 17(t-1)^2 + 4(t-1)^3 \qquad = 21 + 34(t-1) + 12(t-1)^2$$

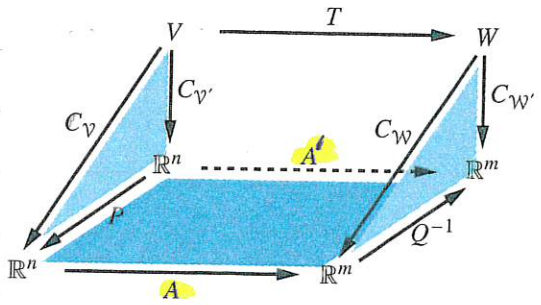
$$\mathcal{V}' = \{1, t-1, (t-1)^2, (t-1)^3\}, \quad C_{\mathcal{V}'}(f) = \begin{bmatrix} 10 \\ 21 \\ 17 \\ 4 \end{bmatrix} \xrightarrow{A'} \begin{bmatrix} 21 \\ 34 \\ 12 \end{bmatrix} = C_{\mathcal{W}'}(f') \quad \mathcal{W}' = \{1, t-1, (t-1)^2\}$$

$$P = \begin{bmatrix} 1 & -1 & 1 & -1 \\ 0 & 1 & -2 & 3 \\ 0 & 0 & 1 & -3 \\ 0 & 0 & 0 & 1 \end{bmatrix} \qquad Q = \begin{bmatrix} 1 & -1 & 1 \\ 0 & 1 & -2 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\mathcal{V} = \{1, t, t^2, t^3\} \xrightarrow{C_{\mathcal{V}}} C_{\mathcal{V}}(f) = \begin{bmatrix} 2 \\ -1 \\ 5 \\ 4 \end{bmatrix} \xrightarrow{A} \begin{bmatrix} -1 \\ 10 \\ 12 \end{bmatrix} = C_{\mathcal{W}}(f') \quad \mathcal{W} = \{1, t, t^2\}$$

定理  $\begin{cases} P = [C_{\mathcal{V}}(v'_1), \dots, C_{\mathcal{V}}(v'_j), \dots, C_{\mathcal{V}}(v'_n)] \quad (\mathcal{V}' \rightarrow \mathcal{V}) \\ Q = [C_{\mathcal{W}}(w'_1), \dots, C_{\mathcal{W}}(w'_j), \dots, C_{\mathcal{W}}(w'_m)] \quad (\mathcal{W}' \rightarrow \mathcal{W}) \end{cases}$

$$\begin{cases} A = [T]_{\mathcal{V}, \mathcal{W}} \\ A' = [T]_{\mathcal{V}', \mathcal{W}'} \end{cases} \Rightarrow \begin{cases} A' = Q^{-1} A P \\ A = Q A' P^{-1} \end{cases}$$



Example 2  $T: \mathcal{M}_{2 \times 2} \rightarrow \mathcal{M}_{2 \times 2}$   
 $X \mapsto MX, M = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} : \begin{bmatrix} a & b \\ c & d \end{bmatrix} \mapsto \begin{bmatrix} c & d \\ a & b \end{bmatrix} \quad (R_{12})$

$\mathcal{M}_{2 \times 2}$ : 基底

$\cdot \mathcal{V}' = \left\{ \begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ -1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & -1 \end{bmatrix} \right\} \mathbb{R}^4 \xrightarrow{A'} \mathbb{R}^4$   
 $\quad \quad \quad \underbrace{\quad}_{v_1'} \quad \underbrace{\quad}_{v_2'} \quad \underbrace{\quad}_{v_3'} \quad \underbrace{\quad}_{v_4'}$   
 $A' = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & -1 \end{bmatrix}$

$P = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \end{bmatrix}$

$A = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix}$

$Q = P$

$\cdot \mathcal{V} = \left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\} \mathbb{R}^4 \xrightarrow{A} \mathbb{R}^4$   
 $\quad \quad \quad \underbrace{\quad}_{v_1} \quad \underbrace{\quad}_{v_2} \quad \underbrace{\quad}_{v_3} \quad \underbrace{\quad}_{v_4}$

$A' = Q^{-1}AP$